# Computational modeling of macrophage polarization dynamics in skeletal malignancies. An integrated in silico and in vivo approach



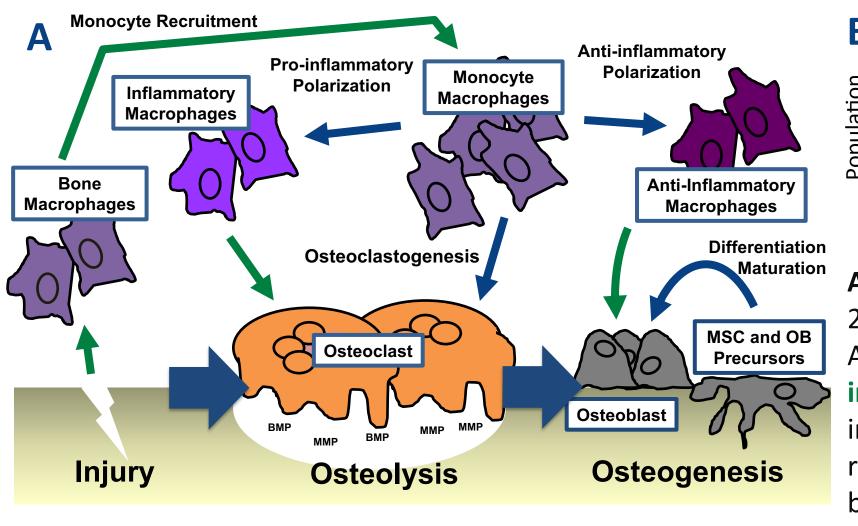
<sup>1,3</sup>Tumor Biology and Integrative Mathematical Oncology Depts., H. Lee Moffitt Cancer Center, 12902 USF Magnolia Drive, Tampa, FL <sup>2</sup>Cancer Biology Ph.D. Program, University of South Florida, Tampa, FL

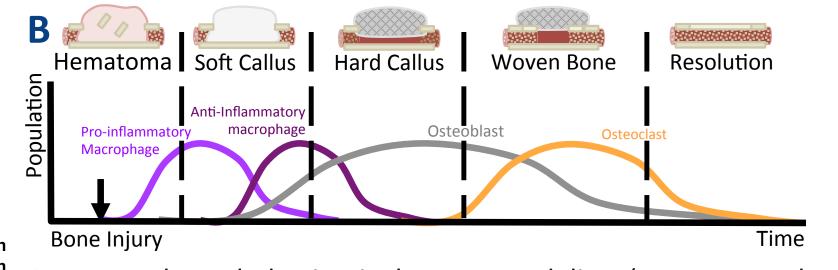


#### 1. Bone-Metastatic Malignancies and The Vicious Cycle

- Various metastatic malignancies are osteophilic, including prostate cancer and breast cancer
- Most cancers are incurable at bone-metastatic stages and cause vicious cycle by disrupting osteolysis and osteogenesis, resulting in poorly-vascularized brittle bone with painful lesions susceptible to fractures
- Macrophage-targeted therapies have enjoyed success in some primary solid malignancies but their application in bone metastatic diseases are unknown
- Here we take an interdisciplinary in silico/in vivo approach to understand the largely unexplored role of macrophages in cancer-bone interaction

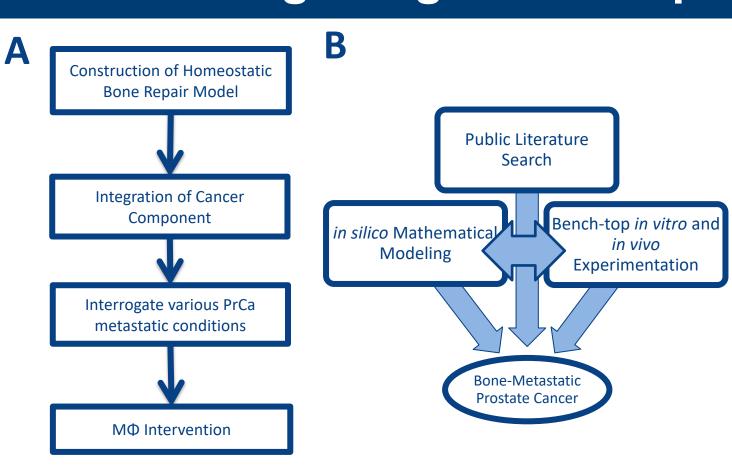
# 2. Macrophages in Bone Remodeling/Repair





A. Macrophage behavior in bone remodeling (Raggatt et al 25285719, Schindeler et al. 18692584 Horwood 26498771). Arrows indicate differentiation process or agonistic interaction. B. Temporally-regulated polarized pro- and antiinflammatory macrophages emerge to trigger clearance and repair responses bye osteoblasts and osteoclasts to restore the bone microenvironment.

#### 3. Integrating Bench Top Experiments and Math Modeling

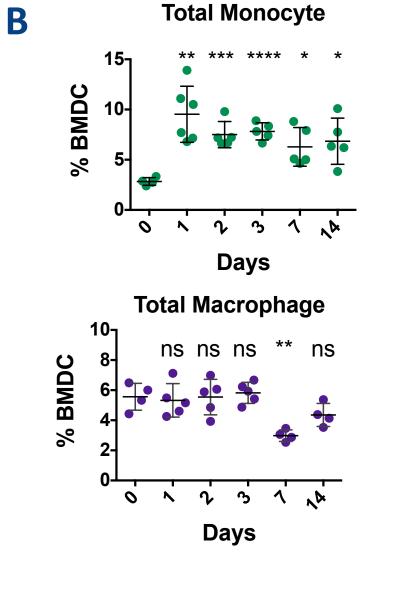


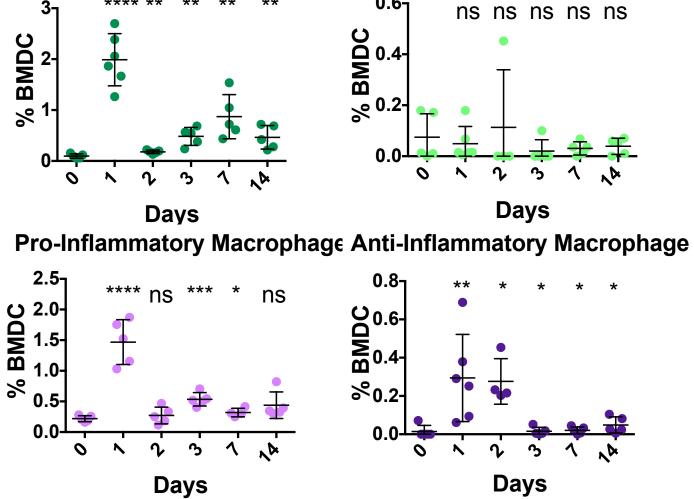
Math modeling facilitates the simultaneous observation of multiple cell populations and their interactions over time within a complex microenvironmental system. Math models powered by empirical and published parameters allow researchers to identify key components within networks of interactions, and make in silico predictions that can be validated biologically. A. The math model was initially parameterized to reflect cellular interactions underlying normal bone healing processes prior to the integration of metastatic prostate cancer cells. **B**. Building in silico models require a blend of empirical and published data to appropriately parameterize the model.

## 4. Macrophages Polarize into Pro- and Anti-Inflammatory States in Response to Bone Injury

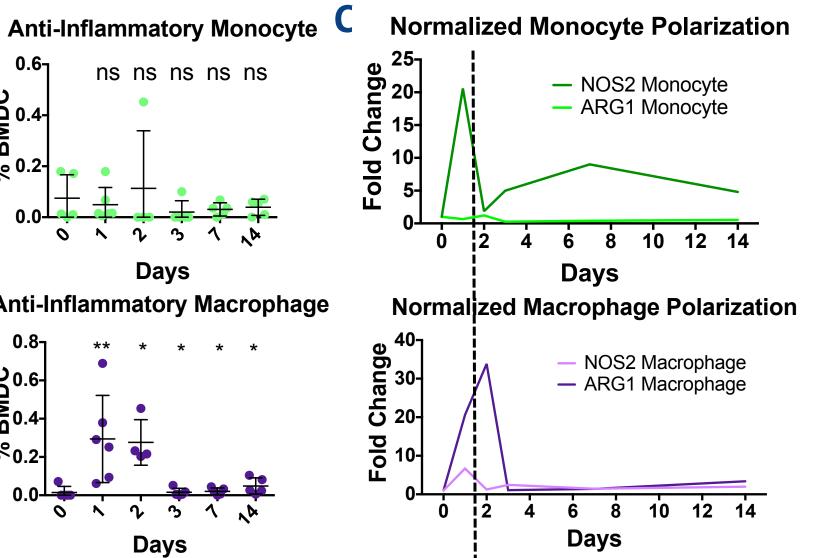
		Marker	Fluorophore
	Phenotyping	CD11b	BV785
		Ly-6G	AF700
		Ly-6C	AF488
		F4/80	BUV395
	Inflammatory Status	NOS2	APC
		ARG-1	PE
		TNF-a	BV650
		IL-4R	BV421
		MHC Class II	PE-Cy7
		Viability	NIR

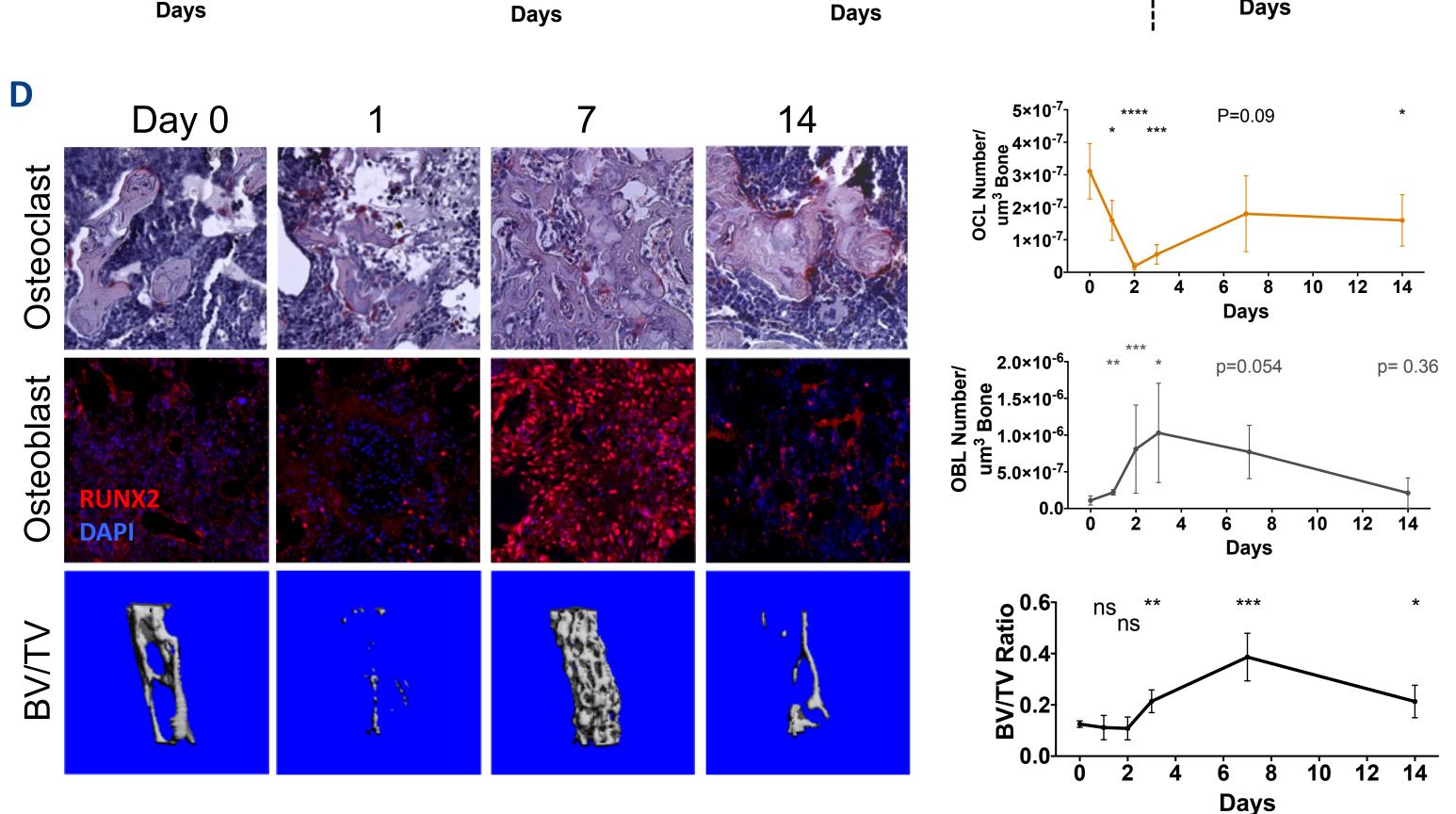
Bone injury was induced by performing mock intratibial injections on C57BL/6 mice (n=30). Sets of 5 mice were taken down for immunologic and histological analyses without injury to establish baseline quantitation and at Day 1, 2, 3, 7 and 14. Bone marrow was flushed from one tibia from each mice for assessing monocyte (CD11b+Ly6C+Ly6G-) macrophage (CD11b+Ly6C-Ly6G-) dynamics and the contralateral tibias were fixed in formalin for histology. A. Panel of marker antibodies were designed to phenotypically dissect myeloid subpopulations and their polarization states. B. Flow Cytometry analysis revealed dynamic shifts in monocyte and macrophage populations and polarization states in response to bone injury. C. Normalized quantitation revealed temporally distinct phases of monocytes and macrophages pro- (NOS2+) and anti-inflammatory (ARG1+) polarization, respectively. D. Fixed bones were subject to uCT analysis to reveal bone volume over time (BV/TV). Decalcified bones were sectioned and subject to osteoclast (TRAcP staining) and osteoblast (RUNX2 immunofluorescence) temporal quantitation.





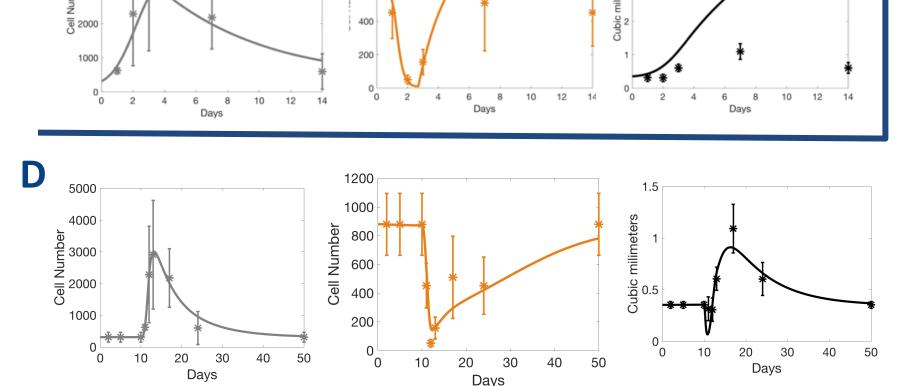
**Pro-Inflammatory Monocyte** 





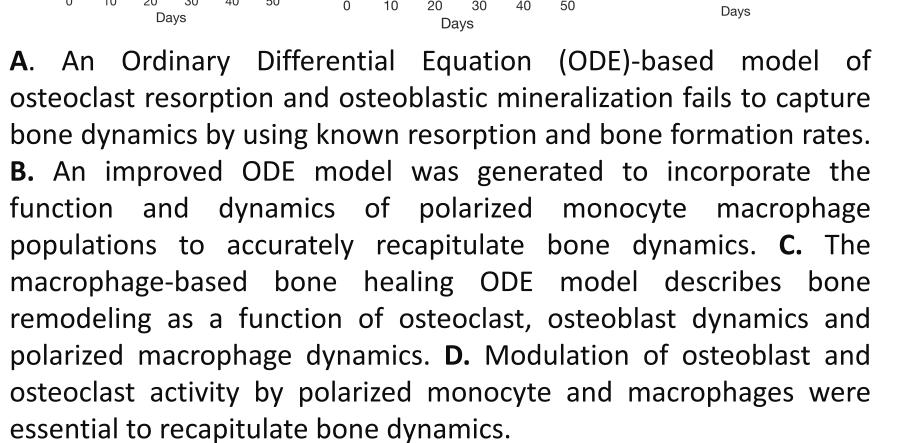
## 5. Monocyte Macrophage Polarization Controls Osteoclast and Osteoblast Activity during Bone Injury Repair

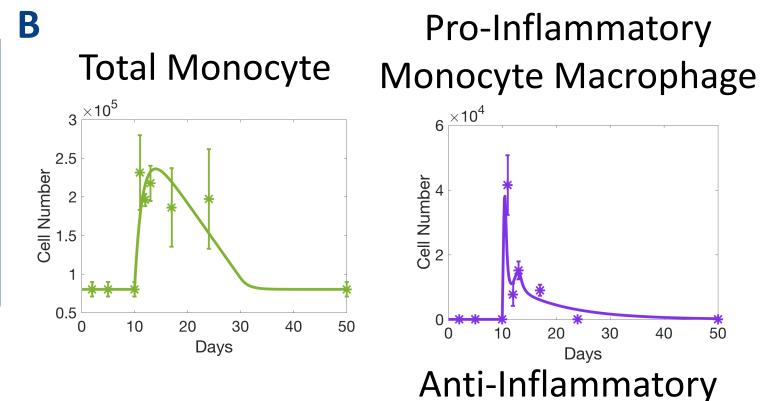
Bone



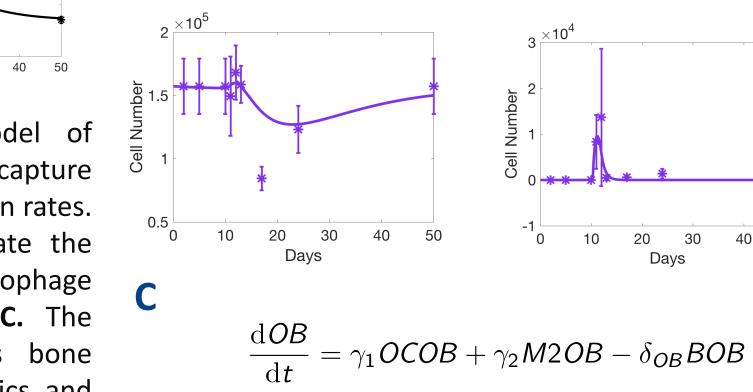
Osteoclast

Osteoblast

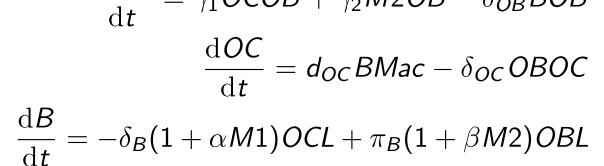




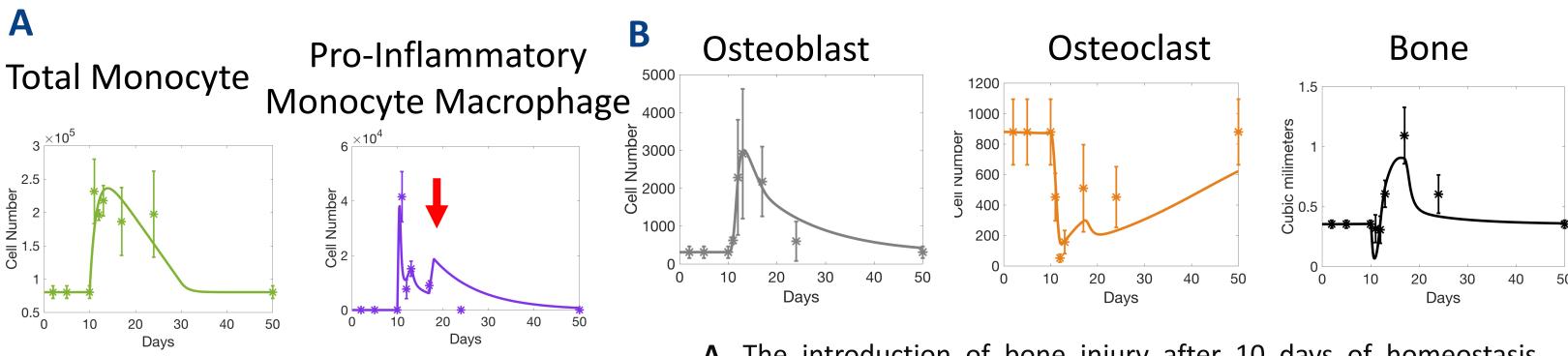
Monocyte Macrophage

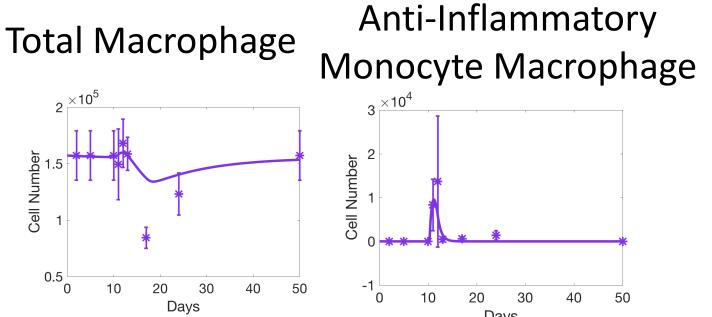


Total Macrophage



#### 6. Modifying Macrophage Polarization Predicts Shorter Bone Healing Times

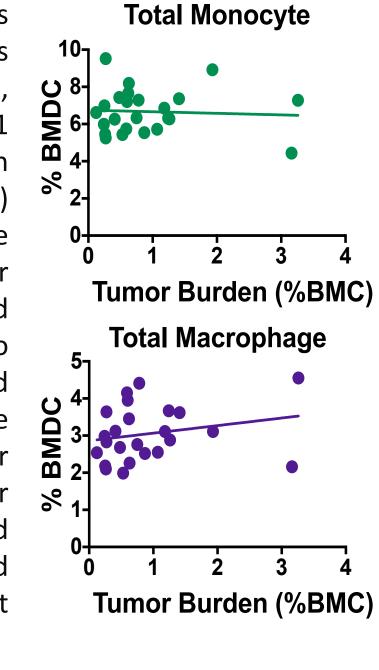


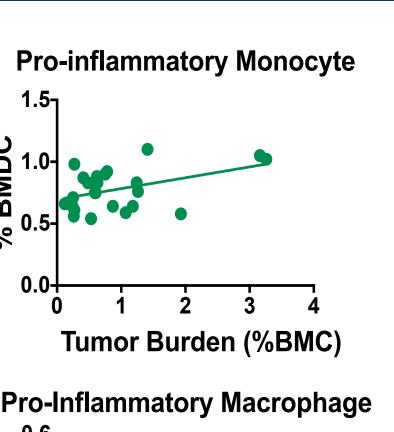


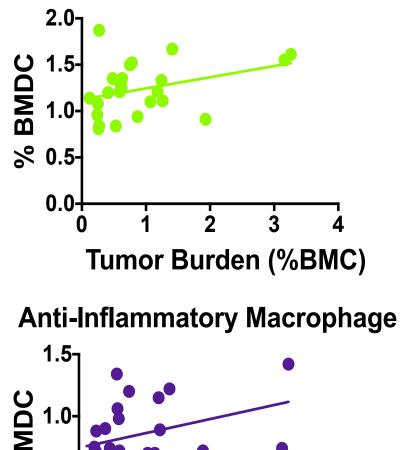
A. The introduction of bone injury after 10 days of homeostasis resulted in expansion of monocytes and shifts in polarized macrophage behavior. Simulation of inflammatory polarization stimulation 7 days following time of injury (arrow) predicted shifts in dynamics. **B.** Modulated pro-inflammatory macrophages subsequently affected osteoblast and osteoclast dynamics and shortened duration of bone repair to reach homeostasis (50 vs 30 days).

# 7. Multiple Myeloma Alters Monocyte and Macrophage Temporal **Dynamics to Generate Osteolytic Disease**

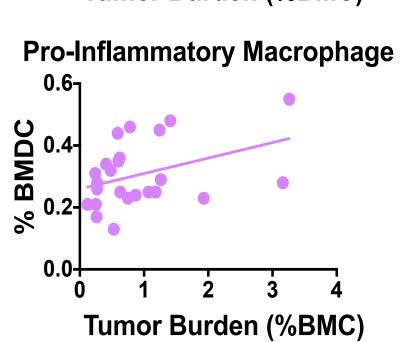
incorporating now macrophage dynamics into models of skeletal malignancy such as Multiple Myeloma. To this end, mice were inoculated with 5TGM1 multiple myeloma cells by tail vein injection (n=20;  $1x10^6$  cells/mice) to establish bone lesions. Mice were sacrificed 7 days after injection and tibias were subjected to multimetric FACS analysis to assess tumor burden (IgG2b+) and macrophage corresponding Linear polarization status. regression curves reveal that tumor burden correlated with pro- and anti-inflammatory monocyte and macrophage polarization but not total populations.

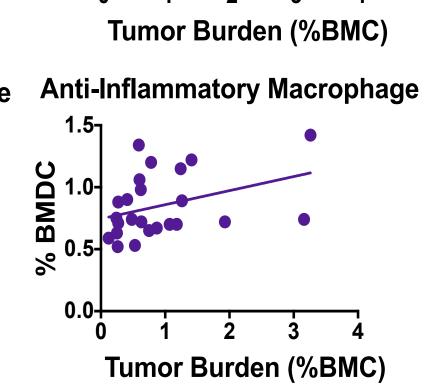






**Anti-Inflammatory Monocyte** 





### 8. Conclusions and Future Directions

- Macrophages are key players in the osteolytic and osteogenic response to bone injury
- The mathematical model, powered by empirical parameters, recapitulate the cellular dynamics of bone injury response
- In vivo injury model revealed persistent myeloid infiltration, and an initial expansion of pro-inflammatory, followed by, antiinflammatory macrophages.
- Mathematical simulation of bone cell dynamics accurately reflected biological dynamics
- Math model can be manipulated to simulate existing therapeutics and offer insight into bone repair progression • Bone-metastatic multiple myeloma generates pro- and anti-inflammatory macrophages over time in vivo
- ODE will be expanded into the cancer context for interrogating strategies to reduce cancer progression and bone pathology

# • Rapid model predictions will assist in understanding tumor-bone biology and guide bench-top therapy design and testing

## 8. Acknowledgments

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